Effects of the automation level on gaze behavior: a full flight simulator campaign with professional airline pilots

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INTRODUCTION

Automation in modern cockpits contributed to improvements in flight safety by reducing pilot workload, fatigue, or increasing situation awareness (Lee & Seppelt, 2012). Yet, whereas lack of automation was problematic in the beginnings of aviation, growing role of automation now raises new challenges with experts pointing at risks associated with an over-reliance of pilots on automatisms. The first risk associated with use of automatisms is the loss of situation awareness associated with pilots being « out-of-the-loop » (Endsley & al., 1995) or unable to effectively monitor or question automated systems when required (Mumaw & al., 2001 ; Parasuraman & al., 1993). Second, when flying with high levels of automation, pilots may be prone to over-confidence (Antonovich, 2008) or automation complacency (Parasuraman & al., 2010) that can result in an improper monitoring of flight instruments that would further challenge pilot abilities to take-over in case of automation failure (Nikolic & Starter, 2007). Improper monitoring has been involved in 80% of major aircraft accidents in the US between 1978-1990 (NTSB, 1994). At last and in the long run, over-relying on automatisms may also induce loss of manual flying skills (Halsbeck & Zhang, 2017).

The objectives of this study were to analyze airline pilots' gaze behavior when using different levels of automation. We hypothesized that gaze behavior would be influenced by the level of automation and pilot's role (pilot-flying or pilot-monitoring) ; that a low level of automation would be associated to lower performances, increased workload and an increased time spent on primary flight parameters ; and that these effects would be more important for pilot-flying.

METHODOLOGY AND RESULTS

Participants

Twenty A320 qualified pilots including 10 Captains and 10 First Officers were recruited to take part in the experiment. All were males, with a mean age of 42 years for Captains and of 29 years for First Officers, and with a flight experience of respectively 11500 flying hours ($SD = 1300$ flying hours) and 3500 flying hours ($SD = 340$ flying hours). All were volunteers, unaware of the purpose of the study, and randomly assigned to another pilot. The experiment was approved by the Air France local committee as well as by the CERNI (Ethics Committee of the University of Toulouse, France, IRB00011835-2020-03-03-210).

Task

All pilots performed three flights, from take-off to landing (with an Instrument Landing System, ILS) at Toulouse airport (LFBO, runway 32R), alternatively as pilot-flying (PF, i.e., pilot actually flying the aircraft) and pilot-monitoring (PM). Weather conditions were standard instrument flying conditions, with a visibility higher than 550m and a 15 knots crosswind. The levels of automation consisted of two systems: Flight Directors (FD) and Autothrust (A/T). Both are Airbus flight guidance systems that are designed to assist the pilot in respectively controlling flight path by providing attitude guidance and aircraft speed by automatically adjusting engines thrust.

For each approach, pilots were instructed to perform the approach in manual flying (i.e., with autopilot disengaged) but with different levels of automation. The three following levels of automation were used:

- Full use of automation: FD ON & A/T OFF
- Partial use of automation: FD ON & A/T OFF
- No use of automation: FD OFF & A/T OFF

Apparatus

Experiments were conducted in a certified A320 Thomson full-flight simulator used for flight crew training. Flight performances data were recorded during the approach including speed and path deviation. Gaze data were recorded using two head mounted Pertech eye-trackers, and five areas-of-interest (AOI) have been considered : window, attitude, speed, engine parameters and path deviation that aggregates heading, lateral deviation scale and vertical deviation scale. Three basic gaze metrics were used to characterize pilot's gaze behavior : the percent time on AOI, the mean glance duration, and the glance rate, that respectively reflect pilot's attention distribution over the different AOIs, effectiveness in information acquisition processes when visiting that AOI, and frequency of visit of that AOI (Halsbeck & Zhang, 2017). Subjective measurements of perceived workload were performed on each level of automation with the NASA-TLX Task Load Index (Hart and Staveland, 1988).

Results

Workload and Flight Performance

As expected, a reduction in the level of automation was associated with a decrease in flight performances and an increase in subjective pilot mental workload.

A decrease in performances was indeed observed in the no-use-of-automation condition (Figure 1), with significantly higher path deviations when pilots did not rely on autothrust nor flight directors. In this condition, 5 pilots out of 20 had to go-around due to being unstabilized during the approach. An increase in subjective workload was also observed with each reduction in level of automation (Figure 2), with a higher subjective workload in the no-use-of-automation condition ($M = 85.9$, $SD = 4.5$) than in the partial-use-of-automation condition ($M = 44$, $SD = 23$) ($t(8) = 5.66$, $p < .001$), and a higher subjective workload in the partial-use-of-automation than in the full-use-of-automation condition ($M = 24$, $SD = 13$) ($t(8) = 5.71$, $p < .001$).
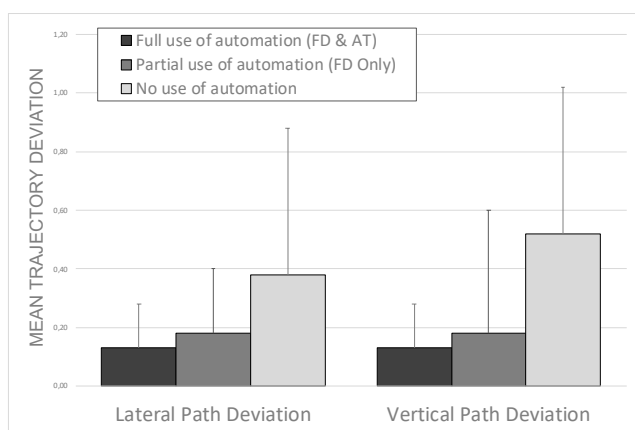

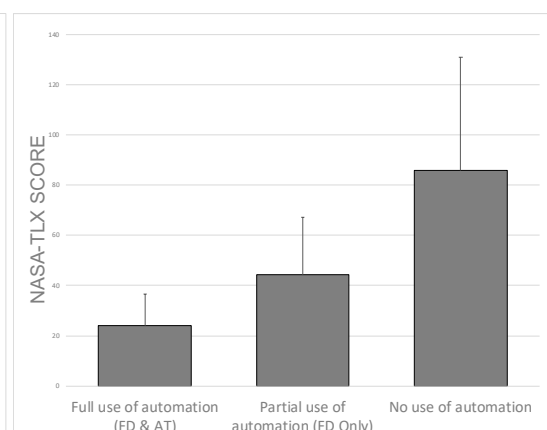Figure 1 – Path Deviations per level of automation

Figure 2 – NASA-TLX Score per level of automation

Influence of the level of automation on PF and PM

Basic Gaze-based metrics

One way (Automation) repeated measures ANOVA were performed on each AOI for percent time on AOI, mean glance duration, and glance rate (Figure 3) to compare PF and PM gaze behavior over the three full-use-of-automation, partial-use-of-automation and no-use-of-automation conditions.

A main effect of Automation was observed for PF on percent time spent on attitude ($F(2,38) = 14.7$, $p < .001$), speed ($F(2,38) = 12.2$, $p < .001$), engine parameters ($F(2,38) = 5.45$, $p = .008$), path deviation ($F(2,38) = 12.5$, $p < .001$) ; on attitude ($F(2,38) = 14.7$, $p < .001$), engine parameters ($F(2,38) = 3.34$, $p < .046$), and path deviation ($F(2,38) = 6.09$, $p = .005$) mean glance duration ; and on engine parameters ($F(2,38) = 3.6$, $p < .037$) and path deviation glance rate ($F(2,38) = 5.99$, $p = .005$). There was no main effect of the level of automation on any of the PM basic gaze metrics, with PM gaze behavior being stable throughout the three levels of automation conditions. Post-hoc comparisons of Automation on PF basic gaze metrics are hereafter presented, with only significant main effects presented in this section ($p < .05$).

When compared to the full-use-of-automation condition, the partial-use-of-automation condition was associated with a significant increase in percent time spent on speed ($t(19) = 4.51$, $p < .001$) and engine parameters ($t(19) = 2.94$, $p = .022$) ; with a significant increase in engine parameters mean glance duration ($t(19) = 3.28$, $p = .011$) ; and with a significant increase in engine parameters glance rate ($t(19) = 2.82$, $p = .028$).
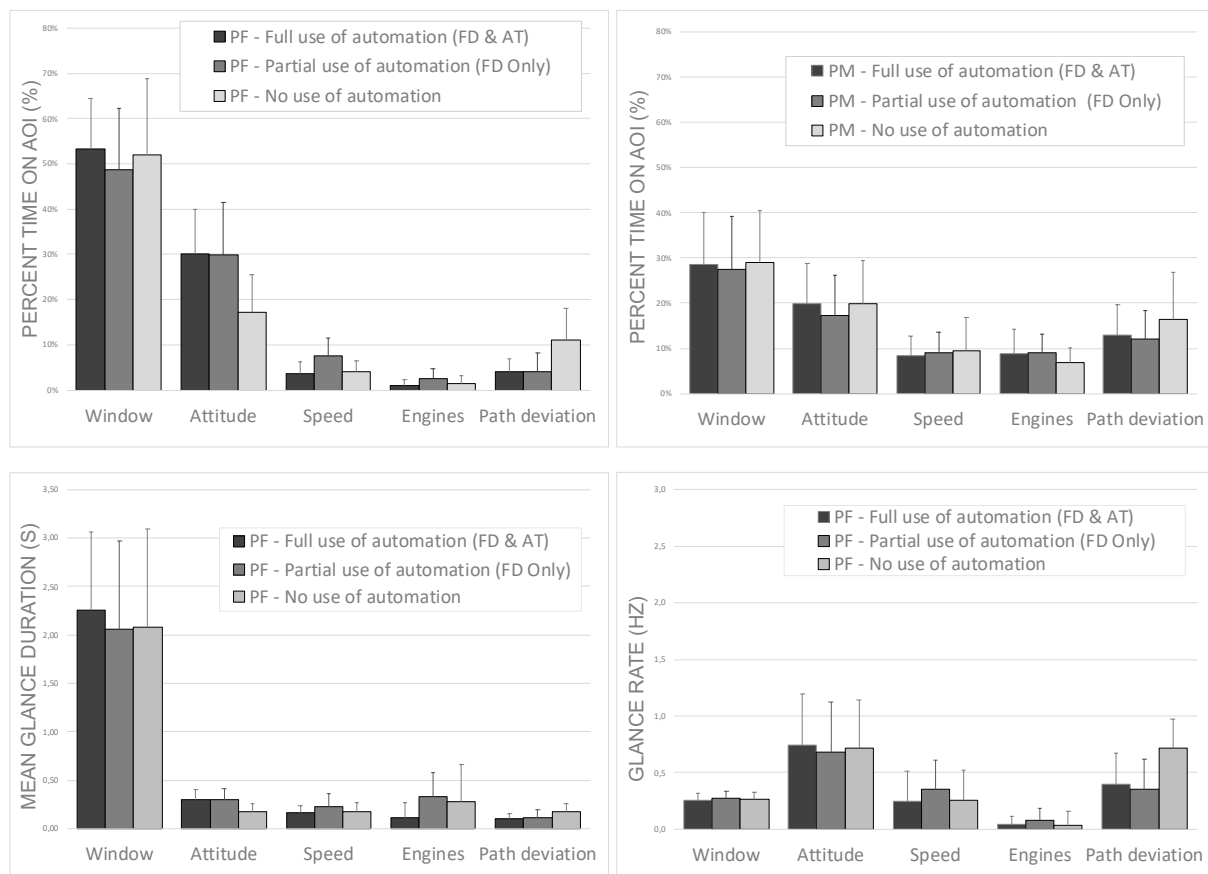

Figure 3 - Basic Gaze Metrics per AOI and per level of automation:
Percent time on AOIs for PF (top-left) and PM (top-right), Mean Glance Duration on AOIs for PF (bottom-left) and Glance Rate for AOIs for PF (bottom-right)

When compared to the partial-use-of-automation condition, the no-use-of-automation condition was associated with a significant reduction in percent time spent on attitude ($t(19) = 4.15$, $p = .002$) and speed ($t(19) = 3.30$, $p = .010$) and a significant increase in percent time spent on path deviation ($t(19) = 3.27$, $p = .011$) ; with a significant reduction in attitude mean glance duration ($t(19) = 4.15$, $p = .002$) and a significant increase in path deviation mean glance duration (($t(19) = 3.71$, $p = .004$) ; and with a significant increase in glance rate on engine parameters ($t(19) = 2.78$, $p = .030$) and path deviation ($t(19) = 2.57$, $p = .047$).

When compared to the full-use-of-automation condition, the no-use-of-automation was associated with a significant reduction in percent time spent on attitude ($t(19) = 4,34$, $p = .001$), with a significant increase in percent time spent on path deviation ($t(19) = 4.60$, $p < .001$) ; with a significant reduction in attitude mean glance duration ($t(19) = 4.34$, $p < .001$) ; and with a significant increase in glance rate on engine parameters ($t(19) = 3.70$, $p = .004$) and path deviation ($t(19) = 2.76$, $p = .032$).

Gaze spatial distribution

We used Static Gaze Entropy (Figure 4) as a measure of gaze spatial distribution over the different AOIs and performed a two way (Role x Automation) repeated measures ANOVA. We found a significant main effect of pilot's role ($F(1,38) = 17,7$, $p < .001$) with pilots exhibiting a more distributed gaze allocation when flying as PM ($M = 2,06$ bits, $SD = 0,11$) than when flying as PF ($M = 1,93$ bits, $SD = 0,16$) ($t(89,67) = 6.04$, $p < .001$). We found no significant main effect of Automation on Static Gaze Entropy ($F(1,76) = 0.75$, $p = .48$). A significant interaction between Automation and Role ($F(2,76) = 3.17$, $p = .047$) was found.
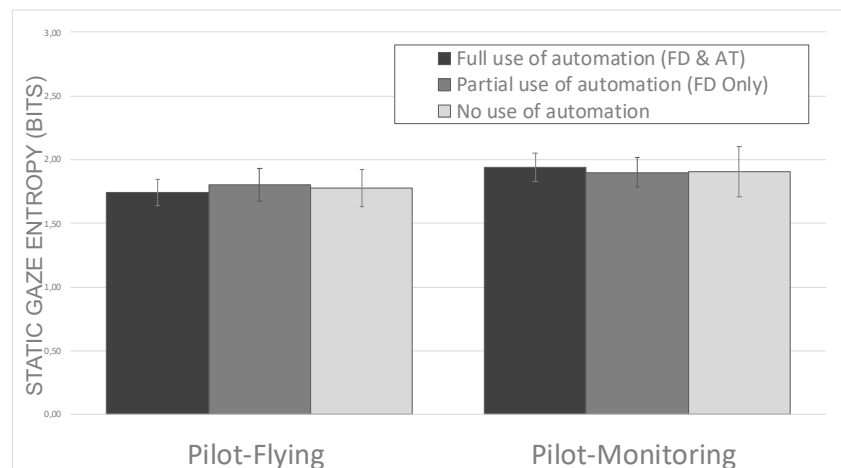

Figure 4 – Static Gaze Entropy as a function of pilot's role and level of automation

DISCUSSION

In this study, we hypothesized that basic gaze metrics would be influenced by the level of automation and by pilots' role as pilot-flying or pilot-monitoring.

Effect of automation on gaze behavior was significant for PFs which is consistent with the fact that the PF is the one actually flying the aircraft. Higher levels of automation were associated with a lower perceived workload and better flight path performances thus emphasizing some beneficial impacts of automation. The reallocation of gaze attention to attitude and flight guidance observed in the highest levels of automation was however at the expense of a more direct monitoring of the flight parameters (speed, engines and path deviation) these automatisms control. Although this shift in attention is a logical consequence of flying with automation, as the pilot delegates speed and path deviation to respectively Flight Directors and

Autothrust, it may reflect a change of reference in pilot's mental modes and representations from flight parameters when flying without automation to flight guidance and automatisms when flying with automation. Such a change could make pilots more vulnerable to losses of situation awareness when flying with automation or unable to regain situation awareness when facing unreliable or inconsistent flight guidance. Whether that behavior is training-induced, training-reversible, task-induced or a consequence of a lower workload or automation complacency is open to question and would justify further eye-tracking based research work.

We observed that PM gaze behavior in terms of basic gaze metrics was generally more spatially distributed over the different AOIs than PFs'. Interestingly, PM gaze behavior was stable across the different levels of automation with PMs therefore maintaining a higher level of direct monitoring of primary flight parameters in the highest levels of automation. Whether this reveals different PF & PM mental modes representations, a lack of adaptation to PF workload, or an absence of need of adaptation, is open to question and points out the relevance for further study of pilot- monitoring gaze behavior. At last, the present study focused on basic gaze metrics that rely on time-averaged data and therefore neglected the information available in the sequence of instrument scanning (Lounis, 2021) thus emphasizing the need for further analysis of the impact of pilot's role and automation on scanpaths.

REFERENCES

Antonovich, B. (2008). The flight crew and automation. *Journal of Aviation/Aerospace Education & Research*, *17*(3), 3._

Ebbatson, M., 2009. The Loss of Manual Flying Skills in Pilots of Highly Automated Airliners (PhD Thesis). Cranfield University, Cranfield.

Endsley, M. R., & Kiris, E. O. (1995). The out-of-the-loop performance problem and level of control in automation. *Human Factors*, *37*(2), 381–394.

Hart, S. G., & Staveland, L. E. (1988). Development of NASA-TLX (Task Load Index): Results of empirical and theoretical research. In P. A. Hancock & N. Meshkati (Eds.), Human mental workload (pp. 139–183). North-Holland.

Haslbeck, A., Zhang, B., 2017. I spy with my little eye: analysis of airline pilots' gaze patterns in a manual instrument flight scenario. Appl. Ergon. 63, 62–71.

Lee, J. D., & Seppelt, B. D. (2012). Human factors and ergonomics in automation design. In G. Salvendy Editor, *Human factors and ergonomics* (Fourth ed., pp. 1615-1642). Hoboken, New Jersey: John Wiley & Sons.

Lefrançois O., Matton N., Causse M., Improving Airline Pilots' Visual Scanning and Manual Flight Performance through Training on Skilled Eye Gaze Strategies. *Safety* 2021,*7*,70.

Lounis C, Peysakhovich V, Causse M (2021) Visual scanning strategies in the cockpit are modulated by pilots' expertise: A flight simulator study. PLoS ONE 16(2): e0247061.

Mumaw, R. J., Sarter, N., & Wickens, C. D. (2001). Analysis of pilots' monitoring and performance on an automated flight deck. *11th International Symposium on Aviation Psychology, Columbus, OH*, *8*.

National Transportation Safety Board. (1994). A review of flightcrew-involved, major accidents of US air carriers, 1978 through 1990. *Safety Study Report No. PB94-97001*.

Nikolic, M. I., & Sarter, N. B. (2007). Flight deck disturbance management: A simulator study of diagnosis and recovery from breakdowns in pilot-automation coordination. *Human Factors*, *49*(4), 553–563.

Parasuraman, R., Molloy, R., & Singh, I. L. (1993). Performance consequences of automation induced'complacency'. *The International Journal of Aviation Psychology*, *3*(1), 1–23.

Parasuraman, R., & Manzey, D. H. (2010). Complacency and bias in human use of automation: An attentional integration. *Human Factors*, *52*(3), 381–410.